

## BLAST OBSERVATIONS OF THE SOUTH ECLIPTIC POLE FIELD: NUMBER COUNTS AND SOURCE CATALOGS

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### ABSTRACT

We present results from a survey carried out by the Balloon-borne Large Aperture Submillimeter Telescope (BLAST) on a  $9\text{ deg}^2$  field near the South Ecliptic Pole at 250, 350, and  $500\text{ }\mu\text{m}$ . The median  $1\sigma$  depths of the maps are 36.0, 26.4, and  $18.4\text{ mJy}$ , respectively. We apply a statistical method to estimate submillimeter galaxy number counts and find that they are in agreement with other measurements made with the same instrument and with the more recent results from *Herschel*/SPIRE. Thanks to the large field observed, the new measurements give additional constraints on the bright end of the counts. We identify 132, 89, and 61 sources with  $S/N \geq 4$  at 250, 350,  $500\text{ }\mu\text{m}$ , respectively and provide a multi-wavelength combined catalog of 232 sources with a significance  $\geq 4\sigma$  in at least one BLAST band. The new BLAST maps and catalogs are available publicly at <http://blastexperiment.info>.

**Key words:** cosmology: observations – galaxies: statistics – methods: data analysis – submillimeter: galaxies

**Online-only material:** color figure, machine-readable tables

### 1. INTRODUCTION

Understanding the formation and evolution of galaxies is one of the foremost goals of experimental cosmology. In the redshift range  $z \simeq 1\text{--}3$ , massive galaxies go through an evolutionary stage characterized by high rates of star formation. These early, dusty galaxies are best characterized by their thermal dust emission at far-infrared and submillimeter wavelengths. They are known to be the main component of the cosmic infrared background (CIB; Fixsen et al. 1998; Hauser & Dwek 2001), but still relatively little is known about their nature and evolution.

Observations by the Balloon-borne Large Aperture Submillimeter Telescope (BLAST; Pascale et al. 2008) have provided the first confusion-limited submillimeter maps at 250, 350, and  $500\text{ }\mu\text{m}$ , with a beam size of  $36''$ ,  $42''$ , and  $60''$ , respectively, covering areas larger than  $1\text{ deg}^2$ . BLAST carried out surveys of the extragalactic sky in two blank fields, one centered on the southern field of the Great Observatories Origins Deep Survey (GOODS-South; Devlin et al. 2009) and one close to the South Ecliptic Pole (SEP). Each survey covers about  $10\text{ deg}^2$ .

BLAST is the first instrument to provide maps of the sky at wavelengths near the peak of the CIB with enough sensitivity, sky coverage, and angular resolution to identify a large number of sources, determine the detailed shape of the source counts, and show that most of the CIB comes from submillimeter

sources already identified in deep  $24\text{ }\mu\text{m}$  surveys (Devlin et al. 2009; Patanchon et al. 2009; Marsden et al. 2009). There have been several studies (Marsden et al. 2009; Pascale et al. 2009) of the statistical properties in the BLAST bands of sources from catalogs defined using *Spitzer*  $24\text{ }\mu\text{m}$  observations (FIDEL; Dickinson et al. 2007). These have shown that  $24\text{ }\mu\text{m}$  sources may well contribute all of the CIB, with about half of the emission coming from galaxies at  $z \gtrsim 1.2$ . Other statistical analyses of the BLAST GOODS-South (BGS) data (Patanchon et al. 2009; Viero et al. 2009) have provided measurements of the differential number counts and of clustering on scales larger than the BLAST beam.

In this paper, we present results from the BLAST survey close to the SEP (hereafter BSEP field). This field was chosen because of very low emission from infrared cirrus, inferred from the IRAS  $100\text{ }\mu\text{m}$  map of Schlegel et al. (1998). Although point-source detection (i.e., at high angular frequency) is not affected by fluctuations at large angular scales (of the order of a degree) from cirrus noise, a low-cirrus region is needed for studies of the CIB at these wavelengths. The SEP field does not have the same richness of multi-wavelength data as GOODS-South, but the multi-wavelength coverage is improving, in particular because of the observations carried out toward that region by the *AKARI* (Matsuhara et al. 2006; Malek et al. 2010; Matsuura et al. 2010), *AzTEC* on the Atacama Submillimeter Telescope

Experiment (Hatsukade et al. 2010), the South Pole Telescope, the Atacama Cosmology Telescope, and the Australia Telescope Compact Array. The catalog presented in this paper can provide submillimeter spectral energy distributions (SEDs) for the AzTEC and AKARI sources. Moreover, the BSEP field is one of the HerMES<sup>15</sup> fields (ADF-S), so that the BSEP catalog can be used for cross-checking of the HerMES ADF-S data.

We have performed an analysis of the statistical properties of the maps to estimate number counts and we have extracted sources using the point-spread function (PSF) to filter the maps, applying the same techniques used in previous BLAST papers (Devlin et al. 2009; Patanchon et al. 2009). A more recent study by Chapin et al. (2010) proposes a different approach for filtering the submillimeter maps, namely a “matched filter” optimized to maximize the signal-to-noise ratio (S/N) of individual point sources in the presence of multiple noise sources. This approach significantly reduces the confusion noise at the expense of slightly higher instrumental white noise. We have also used this approach for the BSEP data and have compared the results using both the PSF and the “matched filter.”

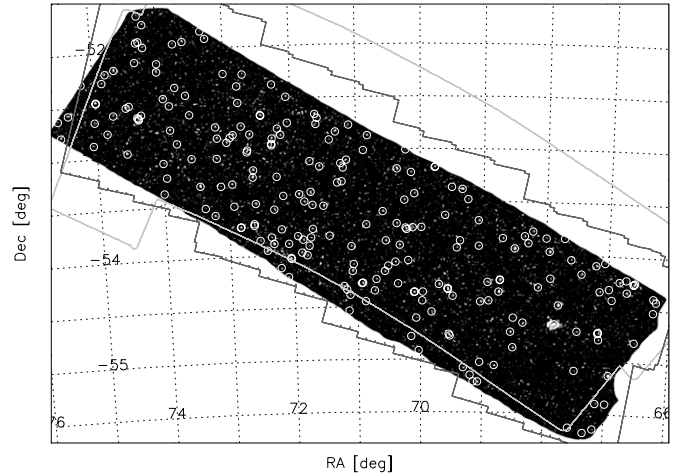
The same field has been also observed by *Spitzer*–MIPS at 24 and 70  $\mu\text{m}$  (Scott et al. 2010). The combination of *Spitzer* and BLAST data enables the identification of mid-IR counterparts for  $\sim 50\%$  of the BLAST sources and also provides useful limits for unidentified sources. The combined results can be used to constrain the SEDs of the galaxies across the entire Rayleigh-Jeans to Wien region of the far-IR spectrum and will be presented in a future paper.

The paper is organized as follows: Section 2 introduces the BLAST data, Section 3 presents the statistical analysis used to calculate differential number counts and a comparison between number counts calculated in this field with previous measurements, as well as comparisons with the most recent evolutionary models, Section 4 provides source lists constructed independently for each BLAST band and combined together, and Section 5 gives some brief conclusions. The source catalogs are provided in the Appendix. This paper constitutes the public release of the BLAST maps and multi-wavelength catalogs of the BSEP field.

## 2. BLAST DATA

During the BLAST 2006 Antarctic flight, deep observations (68 hr) were carried out over a  $9\text{ deg}^2$  field centered on  $(70^\circ 94, -53^\circ 50)$  near the SEP. The BSEP maps are made from a large number of cross-linked scans, producing a uniform map with median  $1\sigma$  sensitivities (equivalent for detecting point sources) of 36.0 mJy, 26.4 mJy, and 18.4 mJy, for the 250, 350, and 500  $\mu\text{m}$  bands, respectively. Figure 1 shows the combined S/N image of all three BLAST bands. These new maps cover an area comparable to the BGS-Wide map of Devlin et al. (2009). We refer the reader to Pascale et al. (2008) for a more detailed description of the characteristics of BLAST and Truch et al. (2009) for information on calibration and data reduction.

The BLAST time-stream data were reduced using a common pipeline to identify spikes, correct detector time drift, and calibrate data (Pascale et al. 2008; Truch et al. 2009). Maps were generated using the SANEPIC software, which uses a maximum-likelihood algorithm to estimate the optimal solution for the map, as well as producing an associated noise map (Patanchon et al. 2008). Absolute calibration is based on observations of the evolved star VY CMa and is estimated



**Figure 1.** Map showing the combined signal-to-noise ratio of all three BLAST bands in the BSEP field. The circles mark the locations of the 232 sources detected with  $S/N \geq 4$  in at least one band, as listed in Table A4. The contours show the overlapping coverage of the  $11.5\text{ deg}^2$  *Spitzer* survey in this field (dark gray) and the region mapped at 90  $\mu\text{m}$  by AKARI (light gray).

to have an uncertainty of 10% (although strongly correlated between the three bands; see Truch et al. 2009 for details).

While the maps represent the optimal weighting of the data across all spatial scales, the largest scales are less constrained due to various systematic effects. This can produce residual large-scale fluctuating patterns across the map. To suppress these spurious signals, all maps have been filtered to remove large-scale angular frequencies, without affecting the contribution of individual point sources. This corresponds to scales in excess of about 10 arcmin (approximately the size of the detector array projected on the sky). This procedure, already used for the BGS maps (e.g., Devlin et al. 2009), also explicitly sets the mean of each map to zero. The filtering induces negative shadows around the locations of bright sources in the maps, but this effect is fully taken into account in the analyses presented in this paper.

The confusion noise in the map is estimated with the method explained in Marsden et al. (2009). We fit a Gaussian to the distribution of the ratio of map pixel values to the instrumental noise  $\sigma_i$ . In the case of no confusion noise, the standard deviation of this Gaussian is unity by definition; if confusion noise is present, it gives the total noise of the map,  $\sigma_t$ , in units of  $\sigma_i$ . The confusion noise,  $\sigma_c$ , is then derived by subtracting the instrumental noise from the total noise in quadrature,  $\sigma_c = \sqrt{\sigma_t^2 - \sigma_i^2}$ . As for the BGS-Wide map, the BSEP map is dominated by instrumental noise and so, using this criterion, the maps are not confusion limited.

The BLAST maps in this release include three primary data products. The first set contains the “raw” maps produced by SANEPIC. The second set of maps have been filtered spatially using a Wiener filter to whiten structures on scales larger than the instantaneous field of view. We also provide “matched-filtered” maps (see Section 4.2): these maps have been filtered using a “matched filter,” which gives superior performs as maps approach the confusion limit (Chapin et al. 2010).

## 3. NUMBER COUNTS

### 3.1. $P(D)$ Analysis

Number counts are estimated using the same method adopted in Patanchon et al. (2009). In that paper it was shown that, in

<sup>15</sup> <http://hermes.sussex.ac.uk>

**Table 1**  
Best-fit Differential Number Counts

250 $\mu\text{m}$			350 $\mu\text{m}$			500 $\mu\text{m}$		
Node (Jy)	Best Fit ( $\log(\text{deg}^2 \text{ Jy}^{-1})$ )	Marginal ( $\log(\text{deg}^2 \text{ Jy}^{-1})$ )	Node (Jy)	Best Fit ( $\log(\text{deg}^2 \text{ Jy}^{-1})$ )	Marginal ( $\log(\text{deg}^2 \text{ Jy}^{-1})$ )	Node (Jy)	Best Fit ( $\log(\text{deg}^2 \text{ Jy}^{-1})$ )	Marginal ( $\log(\text{deg}^2 \text{ Jy}^{-1})$ )
$10^{-4}$	9.39	$<10.09$	$5 \times 10^{-5}$	11.64	$<12.10$	$2.5 \times 10^{-5}$	12.43	$<12.81$
0.1	3.23	$3.27^{+0.11}_{-0.10}$	0.05	3.86	$3.89^{+0.10}_{-0.11}$	0.03	4.08	$4.24^{+0.13}_{-0.19}$
0.25	1.01	$0.88^{+0.34}_{-0.45}$	0.15	1.16	$0.93^{+0.39}_{-0.55}$	0.08	1.66	$1.24^{+0.66}_{-1.14}$
0.5	0.65	$0.75^{+0.21}_{-0.26}$	0.5	0.31	$0.51^{+0.23}_{-0.28}$	0.25	0.58	$0.67^{+0.32}_{-0.38}$
1.2	-0.080	$-0.20^{+0.22}_{-0.25}$	5	-4.09	$< -2.24$	2.5	-11.17	$< -4.27$
10	-14.42	$< -6.78$						

the S/N regime of BLAST, a statistical analysis of the maps is a better approach for obtaining number counts than counting individual sources. This is because it naturally allows for the correction of strong biases due to confusion and flux boosting. This technique also has the advantage of providing an unbiased estimate of the counts at flux densities well below the limit at which sources can be detected individually. The method has been optimized to deal with inhomogeneous noise across the map and filtering to suppress large-scale noise.

We start with a measure of the “probability of deflection,”  $P(D)$ , the histogram of pixel values, and use this to try to obtain the best estimate of the number counts. As in Patanchon et al. (2009), we choose to parameterize the differential number counts by a set of amplitudes at a few predefined fluxes, with the intervals between flux nodes interpolated with power laws to impose continuity of the counts. The number and locations of the nodes are chosen as a compromise between increasing the number of free parameters, to give a better representation of the true underlying counts, and keeping the number of nodes small enough to provide useful estimates of their values. We find that no more than about four amplitude parameters can be estimated for each waveband. This is lower than the numbers estimated from the BGS maps (up to seven nodes at 250  $\mu\text{m}$ ), but this is not unexpected given the extra dynamic range of BGS-Wide plus BGS-Deep. Our observations provide useful information starting from  $\sim 0.05$  Jy and brighter. A larger number of nodes would increase the correlation between neighboring nodes and thus would not add any information.

The  $P(D)$  analysis is carried out by minimizing the negative log-likelihood

$$\Phi(\theta) = - \sum_i n_i \log(p_i(\theta)) - \log(N!) + \sum_i \log(n_i!), \quad (1)$$

where  $n_i$  is the number of pixels with flux densities in the  $i$ th flux bin interval;  $p_i$  is the result of the integral of the  $P(D)$  in the  $i$ th bin, normalized such that  $\sum p_i = 1$ ;  $N$  is the total number of measurements (pixels in the map);  $\theta$  are the parameters of the number counts model and the last two terms derive from the normalization of the multi-nomial distribution function (see Patanchon et al. 2009 for a derivation). Equation (1) is only strictly correct if the probability distribution is the same for all observations (pixels). This is not the case for the BGS maps, with the expected distributions being different for the deep and the wide regions, and it is not completely true for the BSEP maps, even though differences in the noise levels are not as dramatic as in the BGS case. To deal with this we have divided the observed maps into a small number of regions (4), such that in each zone the noise variance is approximately constant.

We then compute the quantities in Equation (1) for each of the regions.

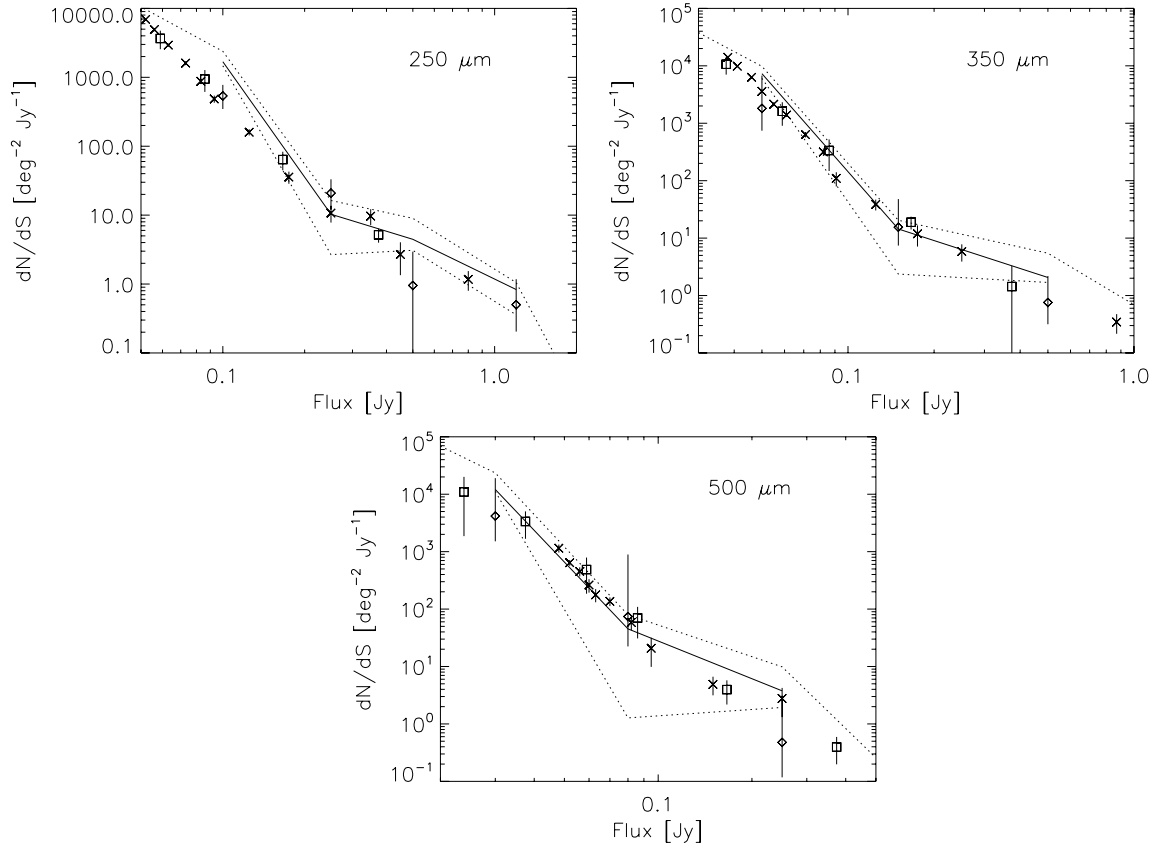
The  $P(D)$  analysis works under the hypothesis that all the sources are point sources, i.e., all sources in the map have the size of the beam. This is not exactly true in our BSEP maps, where at least one local galaxy is partly resolved by the instrument and appears as an extended source (see Section 4.3). Before performing the fit, we have masked the extended source and its dark ring.

We have chosen to fit power laws for differential number counts within predefined flux density bins, as described above. Five distinct power laws are estimated (a total of six free parameters) for the differential number counts at 250  $\mu\text{m}$  and four power laws (five parameters) at both 350 and 500  $\mu\text{m}$ . The choice of flux densities for the boundary nodes is set by requiring them to be very far from the typical values constrained by BLAST, so that the result is independent of our particular choice. Best-fit number counts for the three wavelengths are presented in Figure 2 and Table 1. The uncertainties shown in the figure are computed from the 68% confidence intervals on the marginalized distributions for each parameter separately. The marginal distributions (see Figure 3) have been estimated by sampling the likelihood around its maximum using a Markov Chain Monte Carlo Metropolis Hastings method (MCMCMH; Chib & Greenberg 1995). The median values presented in Table 1 are not exactly the parameters of the best-fit model, due to non-Gaussian likelihoods around the maximum. Pearson correlation matrices for the parameters are given in Tables 2–4.

We compare the predicted histograms of the best-fit multi-power-law model with the actual histograms of the maps in Figure 4. We plot the histograms of the two zones that include about  $\sim 95\%$  of the pixels and ignore the others here, since they only give weak additional constraints on the parameters.

Patanchon et al. (2009) have demonstrated that, in a noise-dominated regime, it is useful to cross-correlate the maps with the beam kernel before  $P(D)$  analysis, even if this might not be the optimal choice. We have performed the analysis both convolving the maps with the same instrumental beam already used for the BGS maps and using a “matched filter,” optimized for confusion-limited maps (see Section 4.2). As already pointed out in Section 2, the BSEP maps are dominated by instrumental noise, thus the “matched filter” does not significantly improve the analysis: the differences between the two methods are much smaller than the uncertainties on the counts.

The technique used in this paper also assumes that galaxies are randomly and independently distributed over the sky. However, significant correlations have been found in the background of the BGS maps (Viero et al. 2009), mainly due to clustering on scales larger than the BLAST beam. As already discussed in Patanchon et al. (2009), the influence of clustering on the measurements



**Figure 2.** Best-fit differential number counts at the three BLAST wavelengths (solid lines). Uncertainties are derived from the marginalized distribution for each parameter (dotted lines); because of the non-Gaussian behavior of the likelihood around its maximum, the best-fit model is not centered on the error bars. The first and last power laws (dotted lines) are upper limits. Number counts in the same bands from other studies are also plotted: Patanchon et al. (2009, diamonds); Oliver et al. (2010, squares); Clements et al. (2010, crosses).

of number counts is negligible with respect to the uncertainties. The clustering in BSEP is comparable to what is measured in BGS, despite having the outskirts of a local cluster (DRCG 0428–53 at  $z = 0.04$ , Dressel 1980) cover almost half of the map. As a test, we repeated the analysis on a sub-map that is free of cluster members. Our previous study on a galaxy cluster suggests that the contribution of the cluster to the submillimeter number counts may be significant in particular at the faint-end flux node in BSEP (Braglia et al. 2010), but the counts derived from the sub-map are consistent with those derived from the full map at the faint end of the distribution. At the bright end the counts from the sub-map are lower, but consistent within the small number statistics. In conclusion, the influence of the cluster can be considered negligible.

### 3.2. Comparison with Other Measurements

The number counts provided in this work can be compared with previously published counts in the same bands. All the existing differential number counts come from very recent maps provided by BLAST and *Herschel*/SPIRE. In Figure 2, we compare our best-fit estimate for the differential number counts with the published counts by Patanchon et al. (2009), Oliver et al. (2010), and Clements et al. (2010). While the Patanchon et al. (2009) counts are derived from a statistical analysis of the map, Oliver et al. (2010) and Clements et al. (2010) extract candidate sources and apply corrections to the raw number counts for Eddington bias, reliability, and incompleteness. Despite these differences, the number counts from these studies are in very good agreement with our results.

The sky coverage delivered by the BSEP map provides further constraints on the number counts at the bright end of the distribution.

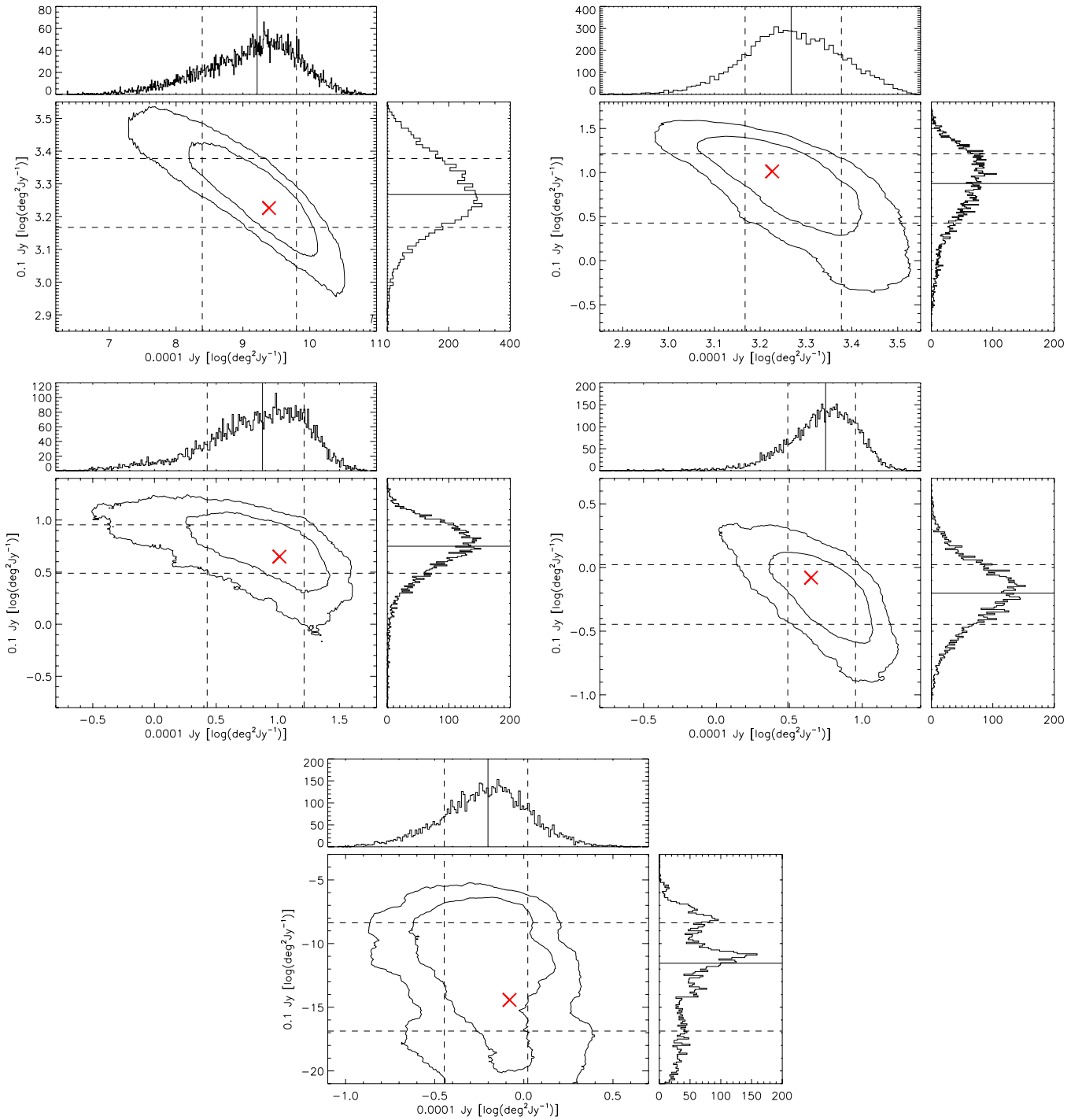
### 3.3. Comparison with Models

Figure 5 compares our best-fit counts with the model of Valiante et al. (2009). As pointed out by Oliver et al. (2010), this model has the distinct advantage, compared with most published models, of being able to fit the rise in the counts for fluxes in the range 10–100 mJy. This could suggest that galaxies at high redshift have on average a colder temperature than local objects with the same luminosity. However, with suitable changes in the luminosity functions, it may be possible to make other models fit the observed counts, so it is premature to conclude that the SEDs need revision. Additionally, the model is able to reproduce the measured counts within the error bars in the range 50–1000 mJy, as explored in these new BSEP data.

## 4. SOURCE CATALOGS

We have compiled a catalog of point sources with flux  $S \geq 3\sigma$  for each band (see Tables A1–A3) using a source-finding algorithm which selects the peaks in a smoothed map produced by the noise-weighted convolution of the image with the telescope PSF (Devlin et al. 2009). The adopted  $\sigma$  is the total noise of the map,  $\sigma_t$ , defined in Section 2. The source lists were synthesized into a common catalog using a procedure which accounts for the significance and positional uncertainty of the counterparts in each band. The radius of the  $1\sigma$  positional error





**Figure 3.** Likelihood distributions and contours for pairs of parameters associated with adjacent nodes at  $250\ \mu\text{m}$ . The two curves in each panel represent 68% and 95% intervals and the solid and dashed lines represent the median and  $1\sigma$  dispersion for each marginalized parameter: the distributions of the parameters are estimated from sampling the likelihood with MCMCMH. The red crosses mark the best-fit values.

(A color version of this figure is available in the online journal.)

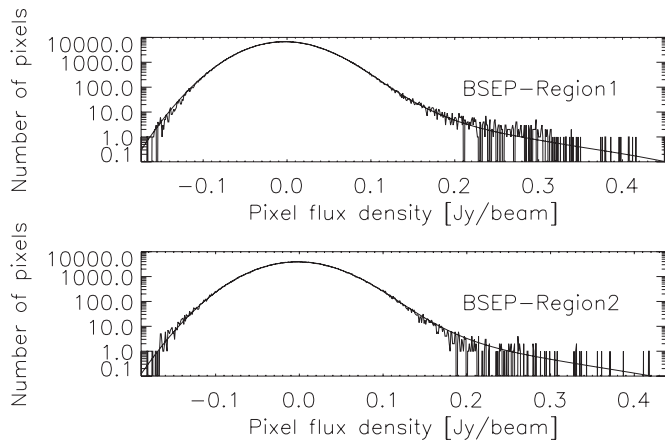
circle,  $\sigma_p$ , for a submillimeter galaxy in a catalog with signal to noise  $\mu$  which has not been corrected for the Eddington bias type of flux boosting can be expressed as

$$\sigma_p = 0.9\theta[\mu^2 - (2\alpha + 4)]^{-1/2} \quad (2)$$

for power-law counts of the form  $N(> S) \propto S^{-\alpha}$ , where  $\theta$  is the full width at half maximum (FWHM) of the telescope beam (Ivison et al. 2007). Using this formula, error circles were calculated assuming  $\alpha = 2$ , as obtained by fitting the

number counts with a single power law (Patanchon et al. 2009). A minimum  $1\sigma$  error circle of  $5''$  was imposed, equal to the  $1\sigma$  pointing uncertainty of the maps.

The combined catalog is comprised of all sources with a significance  $\geq 4\sigma$  in at least one band. Sources from other bands are considered to be matches if they are located within twice the radius of their respective error circles added in quadrature. Positions of sources in the resulting combined catalog were computed by averaging all the positions, weighted by  $\sigma_p^{-2}$ .



**Figure 4.** Histograms of pixel values for two regions having different noise level, including about  $\sim 95\%$  of pixels of the BSEP map, compared with predictions of the best-fit model of the differential counts at  $250\ \mu\text{m}$ . The corresponding plots at  $350$  and  $500\ \mu\text{m}$  look very similar.

The number of sources detected at  $\geq 4\sigma$  by band is 132, 89, and 61 at  $250$ ,  $350$ , and  $500\ \mu\text{m}$ , respectively (see Table 5). Table A4 lists the coordinates and  $250$ ,  $350$ , and  $500\ \mu\text{m}$  flux densities and uncertainties for the 232 sources of the combined catalog.

**Table 2**

Pearson Correlation Matrix for the Parameterized  $dN/dS$  Model at  $250\ \mu\text{m}$

Node (Jy)	$10^{-4}$	0.1	0.25	0.5	1.2	10
$10^{-4}$	1.00	-0.94	0.66	-0.29	-0.002	-0.02
0.1		1.00	-0.79	0.38	-0.09	0.05
0.25			1.00	-0.60	0.21	-0.09
0.5				1.00	-0.71	0.16
1.2					1.00	-0.25
10						1.00

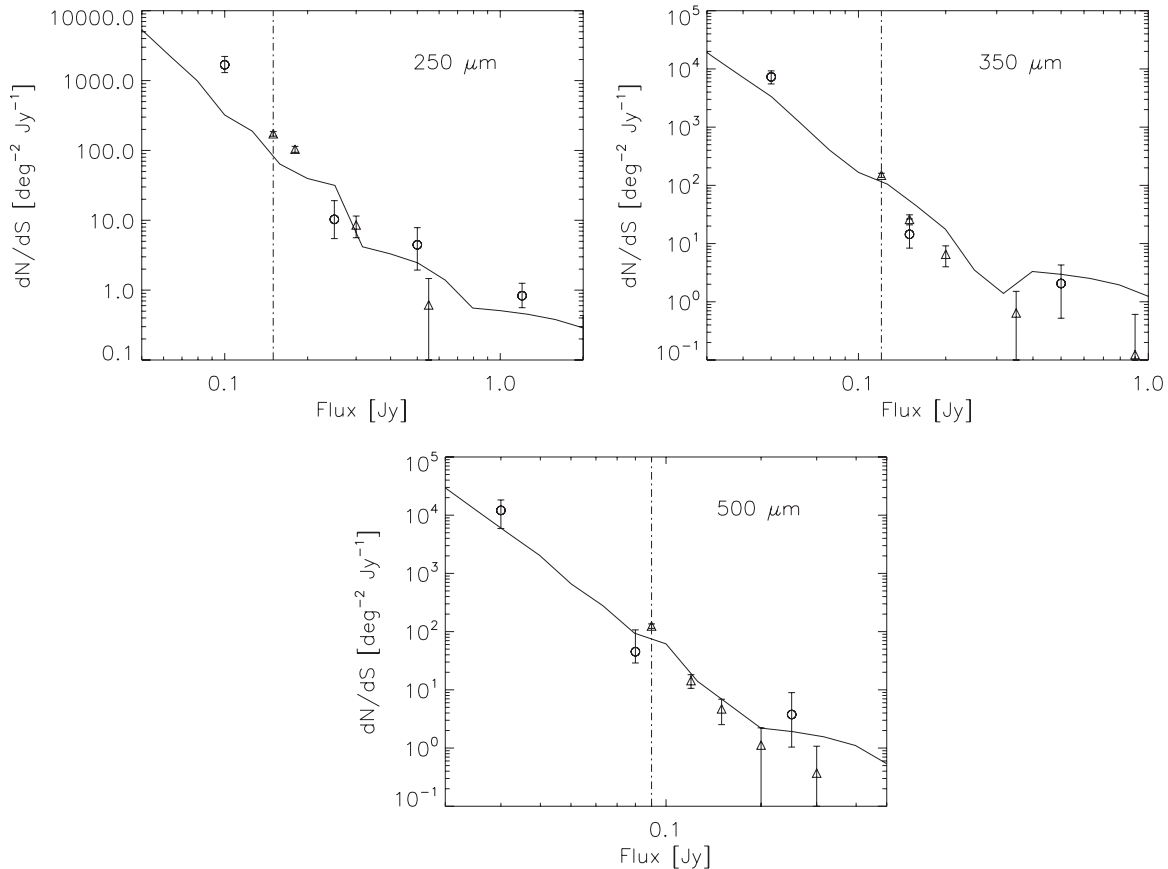
**Notes.** Coefficients are computed as the covariance of the distributions of the parameters around the maximum of the likelihood. The distributions were derived using MCMCMH method.

**Table 3**

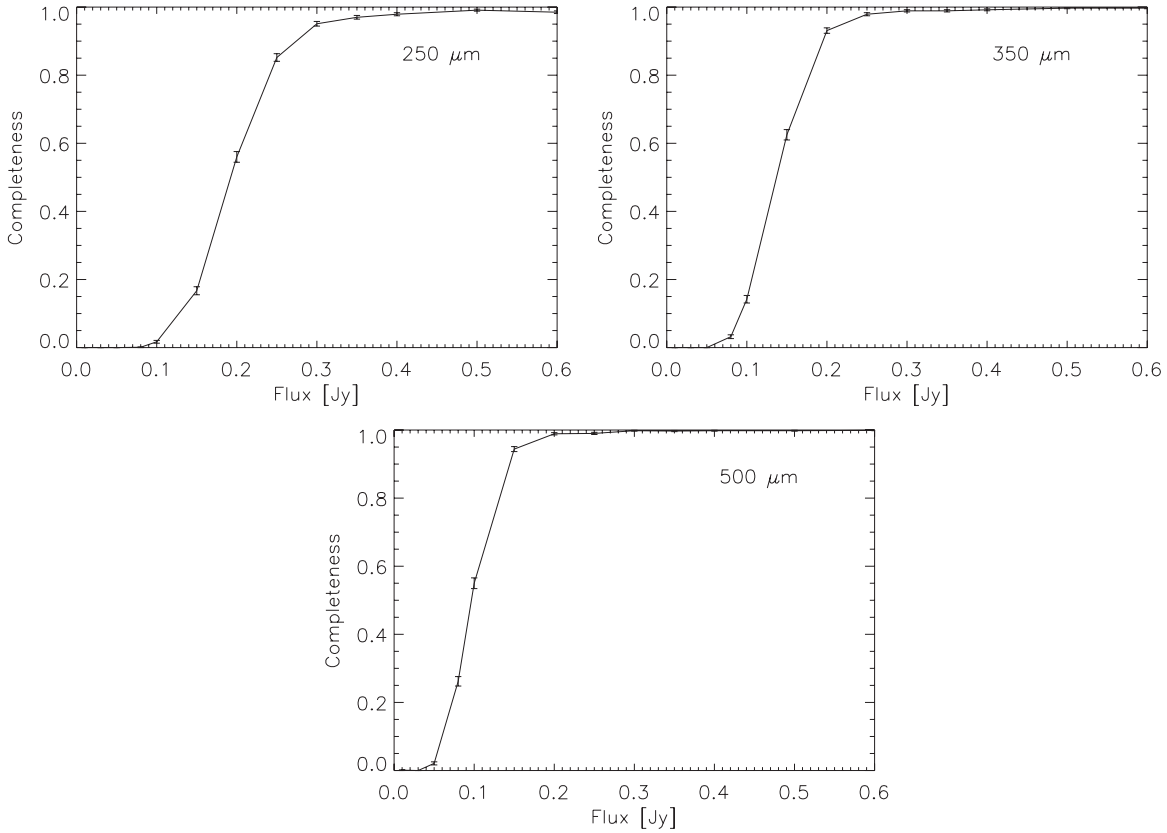
Pearson Correlation Matrix for the Parameterized  $dN/dS$  Model at  $350\ \mu\text{m}$

Node (Jy)	$5 \times 10^{-5}$	0.05	0.15	0.5	5
$5 \times 10^{-5}$	1.00	-0.96	0.63	-0.43	0.07
0.05		1.00	-0.72	0.49	-0.09
0.15			1.00	-0.71	0.16
0.5				1.00	-0.42
5					1.00

**Notes.** Coefficients are computed as the covariance of the distributions of the parameters around the maximum of the likelihood. The distributions were derived using MCMCMH method.



**Figure 5.** Best-fit differential number counts for the three BLAST wavelengths (circles) compared with a realization of the model of Valiante et al. (2009, solid line). The plots also show the cruder estimate of the number counts derived directly from the  $\geq 4\sigma$  catalogs after attempting to correct for completeness and false detection rate (triangles). The vertical lines define the flux limits of the catalogs. Note that the error bars are not marginalized in the same way as for the  $P(D)$  counts, so that it is not possible to make a simple comparison.



**Figure 6.** Catalog completeness at 250, 350, and 500  $\mu\text{m}$ . The error bars are estimated from  $1\sigma$  binomial uncertainties.

**Table 4**

Pearson Correlation Matrix for the Parameterized  $dN/dS$  Model at 500  $\mu\text{m}$

Node (Jy)	$2.5 \times 10^{-5}$	0.03	0.08	0.25	2.5
$2.5 \times 10^{-5}$	1.00	-0.96	0.67	-0.44	0.02
0.03		1.00	-0.73	0.50	-0.01
0.08			1.00	-0.67	0.10
0.25				1.00	-0.26
2.5					1.00

**Notes.** Coefficients are computed as the covariance of the distributions of the parameters around the maximum of the likelihood. The distributions were derived using MCMCMH method.

#### 4.1. Completeness and False Detection Rates

“Completeness” is the probability of detecting a source of a given intrinsic flux density, given the depth of the survey and the source extraction algorithm. We estimate completeness by adding to the initial maps 1000 artificial point sources at random positions and calculating their recovery rates after performing the same source extraction procedure as for the real data. The artificial point sources are modeled as the PSF scaled by the flux density. To avoid blending the simulated sources together or with the existing sources, every simulated source is injected at a distance larger than 2.5 times the half width at half maximum from any real sources and from any other simulated ones. In principle, this procedure introduces a bias which we estimate to be well under 1%. The number density of artificial sources is such that it does not appreciably change the noise properties of the maps. A source was considered to be detected if there was a detection within a circle centered on the source and within the radius of the FWHM of the corresponding wavelength. The catalog completeness as a function of intrinsic flux density

is shown in Figure 6. Table 5 gives the 50%, 80%, and 95% completeness flux densities at each wavelength.

We calculate the false detection rates (FDRs) by performing the extraction algorithm on 100 simulated noise realization maps. This is a conservative overestimate of the FDRs and has the advantage of being model independent (Perera et al. 2008). FDR values are reported in Table 5 for sources with a significance of  $\geq 4\sigma$ . At fluxes  $\geq 5\sigma$  they are all consistent with zero.

Figure 5 shows the number counts derived from the  $\geq 4\sigma$  catalogs after correcting for completeness and false detection rate. We can see that the catalog provides counts in good reasonably agreement with those derived by the  $P(D)$  analysis, but for a restricted flux density range.

#### 4.2. Matched Filter

We have repeated the source extraction procedure along with completeness and false detection rate tests on maps which have been convolved with the “matched filter,” calculated following the algorithm of Chapin et al. (2010) instead of the PSF. This filter is optimized to maximize the S/N of individual point sources in the presence of noise: it significantly reduces the confusion noise and slightly increases the white instrumental noise. This filter should perform significantly better as maps become more dominated by confusion.

Since the BSEP maps are dominated by instrumental noise, we find little benefit in the use of the “matched filter,” as shown in Table 6. Even though we identify a larger number of detected sources using the “matched filter,” the fraction of false detections is also larger. For this reason, we prefer to publish the more conservative, and easily reproducible, source lists derived by filtering the maps with the PSF.

**Table 5**  
BLAST–SEP Maps and Catalogs

Band	$\sigma_i$	$\sigma_c$	Number of Sources	False Detections	50% Completeness	80% Completeness	95% Completeness
( $\mu\text{m}$ )	(mJy)	(mJy)	$> 4\sigma$	$> 4\sigma$	(mJy)	(mJy)	(mJy)
250	36.0	20.6	132	2.28	192	241	299
350	26.4	18.2	89	0.94	137	179	220
500	18.4	15.2	61	0.14	97	132	157

**Table 6**  
BLAST–SEP Maps and Catalogs with Matched Filter

Band	$\sigma_i$	$\sigma_c$	Number of Sources	False Detections	50% Completeness	80% Completeness	95% Completeness
( $\mu\text{m}$ )	(mJy)	(mJy)	$> 4\sigma$	$> 4\sigma$	(mJy)	(mJy)	(mJy)
250	38.2	16.5	145	7.53	194	241	297
350	28.9	14.2	100	6.80	144	184	226
500	20.7	11.6	90	3.13	99	134	163

#### 4.3. Extended Sources

The technique we have used to estimate the flux densities of the BLAST sources is not accurate in the case of extended sources. In the BSEP map there is one clearly extended source, identified with the galaxy NGC 1617 and detected as BLAST J043137–543604 in our catalog. We have performed aperture photometry on this source and have estimated the uncertainty on the flux density taking into account both instrumental noise and calibration uncertainties. The measured values are reported in Table A4.

## 5. CONCLUSIONS

We used BLAST to image a  $9 \text{ deg}^2$  field near the SEP at 250, 350, and  $500 \mu\text{m}$ , achieving median  $1\sigma$  depths of 36.0, 26.4, and 18.4 mJy at each wavelength. We have identified 132, 89, and 61 sources with  $S/N \geq 4\sigma$  at 250, 350, and  $500 \mu\text{m}$ , respectively. These have been compiled into a multi-wavelength catalog of 232 sources with a significance  $\geq 4\sigma$  in at least one BLAST band.

Using the  $P(D)$  technique, we have measured the differential number counts up to 1.2, 0.5 and  $0.25 \text{ Jy}$  at 250, 350, and  $500 \mu\text{m}$ , respectively. The new measurements agree with previous results from BLAST and more recent results from SPIRE

and, thanks to the large area observed, give improved constraints at the bright end of the counts.

We have released the BLAST maps and catalogs to the public at <http://blastexperiment.info>.

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## APPENDIX

### DATA TABLES

The following data tables are provided in this appendix: the lists of submillimeter sources with flux  $\geq 3\sigma$  at  $250 \mu\text{m}$  (Table A1),  $350 \mu\text{m}$  (Table A2), and  $500 \mu\text{m}$  (Table A3); the multi-wavelength combined catalog of sources with a significance  $\geq 4\sigma$  in at least one band (Table A4). Data tables are published in their entirety in the electronic edition of the journal. The first ten entries of Tables A1–A4 are shown here for guidance regarding their form and content.

**Table A1**  
Catalog of BLAST  $250 \mu\text{m}$  Sources in SEP with Significance  $\geq 3\sigma$

ID	BLAST ID	R.A. (J2000)	Decl. (J2000)	$S_{250}$ (Jy)	$\sigma_{250}$ (Jy)	S/N
1	BLAST250 J043138–543604	67.909134	–54.601269	2.05 <sup>a</sup>	0.21 <sup>a</sup>	9.76
2	BLAST250 J043516–541902	68.818535	–54.317352	0.818	0.041	19.77
3	BLAST250 J043126–542507	67.860474	–54.418732	0.698	0.043	16.15
4	BLAST250 J043422–535358	68.591782	–53.899590	0.549	0.041	13.40
5	BLAST250 J044936–535427	72.400566	–53.907761	0.515	0.040	12.78
6	BLAST250 J042755–550332	66.980072	–55.058910	0.716	0.053	13.61
7	BLAST250 J045412–532127	73.550888	–53.357536	0.482	0.041	11.78
8	BLAST250 J045053–531233	72.724892	–53.209232	0.482	0.042	11.56
9	BLAST250 J043424–544132	68.601288	–54.692341	0.478	0.042	11.44
10	BLAST250 J044801–532609	72.005684	–53.435989	0.467	0.043	10.97

**Notes.** Flux densities are not corrected for Eddington-type bias.

<sup>a</sup> These flux densities come from aperture photometry. See Section 4.3 for details.

(This table is available in its entirety in a machine-readable form in the online journal. A portion is shown here for guidance regarding its form and content.)



**Table A2**  
Catalog of BLAST 350  $\mu\text{m}$  Sources in SEP with Significance  $\geq 3\sigma$

ID	BLAST ID	R.A. (J2000)	Decl. (J2000)	$S_{350}$ (Jy)	$\sigma_{350}$ (Jy)	S/N
1	BLAST350 J043137–543603	67.904358	–54.601097	1.09 <sup>a</sup>	0.12 <sup>a</sup>	9.08
2	BLAST350 J043516–541902	68.818398	–54.317253	0.351	0.032	10.92
3	BLAST350 J042754–550331	66.975388	–55.058666	0.472	0.044	10.80
4	BLAST350 J043125–542516	67.855537	–54.421322	0.325	0.033	9.88
5	BLAST350 J043421–535408	68.591515	–53.902260	0.280	0.031	8.93
6	BLAST350 J045412–532127	73.550613	–53.357510	0.272	0.031	8.66
7	BLAST350 J044935–535437	72.396095	–53.910458	0.248	0.032	7.80
8	BLAST350 J045044–533143	72.687477	–53.528831	0.229	0.031	7.37
9	BLAST350 J044706–524018	71.777740	–52.671692	0.204	0.031	6.60
10	BLAST350 J043424–544132	68.601440	–54.692429	0.198	0.032	6.11

**Notes.** Flux densities are not corrected for Eddington-type bias.

<sup>a</sup> These flux densities come from aperture photometry. See Section 4.3 for details.

(This table is available in its entirety in a machine-readable form in the online journal. A portion is shown here for guidance regarding its form and content.)

**Table A3**  
Catalog of BLAST 500  $\mu\text{m}$  Sources in SEP with Significance  $\geq 3\sigma$

ID	BLAST ID	R.A. (J2000)	Decl. (J2000)	$S_{500}$ (Jy)	$\sigma_{500}$ (Jy)	S/N
1	BLAST500 J043137–543603	67.904343	–54.601093	0.395 <sup>a</sup>	0.055 <sup>a</sup>	7.18
2	BLAST500 J042756–550322	66.985207	–55.056328	0.257	0.030	8.52
3	BLAST500 J043125–542506	67.855965	–54.418583	0.167	0.025	6.73
4	BLAST500 J045056–531633	72.735146	–53.275887	0.150	0.023	6.40
5	BLAST500 J043516–541912	68.818130	–54.320080	0.147	0.024	6.21
6	BLAST500 J043521–550056	68.840088	–55.015621	0.147	0.026	5.73
7	BLAST500 J044742–533019	71.926003	–53.505398	0.130	0.024	5.36
8	BLAST500 J045600–524902	74.001236	–52.817482	0.115	0.023	5.04
9	BLAST500 J042640–541057	66.669762	–54.182514	0.150	0.027	5.56
10	BLAST500 J045046–533133	72.692162	–53.526104	0.114	0.023	5.00

**Notes.** Flux densities are not corrected for Eddington-type bias.

<sup>a</sup> These flux densities come from aperture photometry. See Section 4.3 for details.

(This table is available in its entirety in a machine-readable form in the online journal. A portion is shown here for guidance regarding its form and content.)

**Table A4**  
Multi-wavelength Catalog of BLAST Sources in SEP with  $\geq 4\sigma$  Detection in at least One Band

ID	BLAST ID	R.A. (J2000)	Decl. (J2000)	$S_{250}$ (Jy)	$\sigma_{250}$ (Jy)	$S_{350}$ (Jy)	$\sigma_{350}$ (Jy)	$S_{500}$ (Jy)	$\sigma_{500}$ (Jy)
1	BLAST J043137–543604	67.905945	–54.601154	2.05 <sup>a</sup>	0.21 <sup>a</sup>	1.09 <sup>a</sup>	0.12 <sup>a</sup>	0.395 <sup>a</sup>	0.055 <sup>a</sup>
2	BLAST J043516–541904	68.818398	–54.317852	0.818	0.041	0.351	0.032	0.147	0.024
3	BLAST J043125–542510	67.857567	–54.419716	0.698	0.043	0.325	0.033	0.167	0.025
4	BLAST J043422–535403	68.591995	–53.901043	0.549	0.041	0.280	0.031	0.094	0.023
5	BLAST J044935–535432	72.398254	–53.909065	0.515	0.040	0.248	0.032	0.070	0.023
6	BLAST J042755–550329	66.979797	–55.058105	0.716	0.053	0.472	0.044	0.257	0.030
7	BLAST J045412–532127	73.550751	–53.357513	0.482	0.041	0.272	0.031	0.092	0.024
8	BLAST J045053–531232	72.724510	–53.208916	0.482	0.042	0.197	0.033	0.085	0.024
9	BLAST J043424–544132	68.601357	–54.692379	0.478	0.042	0.198	0.032	< 0.074	–
10	BLAST J044801–532609	72.007538	–53.435932	0.467	0.043	0.172	0.033	0.097	0.024

**Notes.** Flux densities are not corrected for Eddington-type bias. The non-detections are expressed as  $3\sigma$  upper limits.

<sup>a</sup> These flux densities come from aperture photometry. See Section 4.3 for details.

(This table is available in its entirety in a machine-readable form in the online journal. A portion is shown here for guidance regarding its form and content.)

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